# MIS780 Advanced AI For Business - Assignment 2 - T2 2024

Task Number: Business Problem Name

Student Name: enter your full name here

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### 1. Executive Summary

Use this section to introduce the business problem, data set, method, experiments, and obtained results

#### 2. Data Preprocessing

Carry out necessary data preprocessing and exploration.

```
import numpy as np
import matplotlib.pyplot as plt
import random
from tensorflow import keras
import tensorflow as tf

tf.config.list_physical_devices('GPU')

from google.colab import drive
drive.mount('/content/drive')

# to show the folders under the dataset
!ls "/content/drive/MyDrive/Colab Notebooks/Part2_WasteImages"
```

```
import os
# Set the paths to the folders containing the image files
cardboard = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Cardboard'
glass = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Glass'
metal = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Metal'
paper = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Paper'
plastic = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Plastic'
vegetation = '/content/drive/MyDrive/Colab
Notebooks/Part2 WasteImages/Vegetation'
# get a list of all files in the folder
cardboard file list = os.listdir(cardboard)
glass file list = os.listdir(glass)
metal file list = os.listdir(metal)
paper file list = os.listdir(paper)
plastic file list = os.listdir(plastic)
vegetation file list = os.listdir(vegetation)
# print the total number of files
print(f'Total number of files under cardboard folder are:
{len(cardboard file list)}')
print(f'Total number of files under glass folder are:
{len(glass file list)}')
print(f'Total number of files under metal folder are:
{len(metal file_list)}')
print(f'Total number of files under paper folder are:
{len(paper file list)}')
print(f'Total number of files under plastic folder are:
{len(plastic file_list)}')
print(f'Total number of files under vegetation folder are:
{len(vegetation file list)}')
import os
import tensorflow as tf
# Create a list to store the image data and labels
data task2 = []
# Iterate through the files in the first folder
for file in os.listdir(cardboard):
  # Check if the file is a jpeg or jpg file
  if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(cardboard, file))
```

```
img = tf.image.decode_jpeg(img,channels=3)
    img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'cardboard'
    # Add the image data and label to the data list
    data task2.append((img, label))
# Iterate through the files in the second folder
for file in os.listdir(glass):
  # Check if the file is a jpeg or jpg file
  if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(glass, file))
    img = tf.image.decode jpeg(img,channels=3)
    img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'qlass'
    # Add the image data and label to the data list
    data task2.append((img, label))
for file in os.listdir(metal):
 # Check if the file is a jpeg or jpg file
  if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(metal, file))
    img = tf.image.decode jpeg(img,channels=3)
    img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'metal'
    # Add the image data and label to the data list
    data task2.append((img, label))
for file in os.listdir(paper):
 # Check if the file is a jpeg or jpg file
  if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(paper, file))
    img = tf.image.decode_jpeg(img,channels=3)
    img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'paper'
    # Add the image data and label to the data list
    data task2.append((img, label))
for file in os.listdir(plastic):
 # Check if the file is a jpeg or jpg file
  if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(plastic, file))
    img = tf.image.decode jpeg(img,channels=3)
```

```
img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'plastic'
    # Add the image data and label to the data list
    data task2.append((img, label))
for file in os.listdir(vegetation):
 # Check if the file is a jpeg or jpg file
 if file.endswith('.jpeg') or file.endswith('.jpg'):
    # Load the image data from the file using TensorFlow
    img = tf.io.read file(os.path.join(vegetation, file))
    img = tf.image.decode jpeg(img,channels=3)
    img = tf.image.resize(img, (50, 50))
    # Assign a label to the file
    label = 'vegetation'
    # Add the image data and label to the data list
    data task2.append((img, label))
# Shuffle the data and split into train/test sets
random.shuffle(data task2)
train data, test data = data task2[:int(len(data task2) * 0.7)],
data task2[int(len(data task2) * 0.3):]
# Extract the image data and labels from the training data
X train, Y train = zip(*train data)
# Extract the image data and labels from the testing data
X test, Y test = zip(*test_data)
# Convert the image data and labels into NumPy arrays
X train = np.array(X train)
Y train = np.array(Y train)
X \text{ test} = np.array(X \text{ test})
Y test = np.array(Y test)
```

#### 3. Predictive Modeling

Create and explain your models (e.g., model architecture, model parameters). Evaluate the models on the experimental data sets.

```
# change integers to 32-bit floating point numbers
X_train = X_train.astype('float32')
X_test = X_test.astype('float32')
# normalize each value for each pixel for the entire vector for each input
X_train /= 255
X_test /= 255
```

```
# print the shape of the reshaped data
print("Training matrix shape", X_train.shape)
print("Testing matrix shape", X test.shape)
import numpy as np
print('The original format of class of the first element in the
training dataset is: ',Y_train[0], '\n')
# Tao môt ma'ng NumPy chứa các nhãn rác
categories = np.array(['cardboard', 'glass', 'metal', 'paper',
'plastic', 'vegetation'])
# Tao một ba'n độ` từ chuộ~i danh mục đế n số nguyên
category map = {
    'cardboard': 0,
    'glass': 1,
    'metal': 2,
    'paper': 3,
    'plastic': 4,
    'vegetation': 5
}
# Mã hóa các danh muc thành số nguyên
Y train = np.array([category map[category] for category in Y train])
Y test = np.array([category map[category] for category in Y test])
print('The unique integer mapping encoding format of the class of the
first element in the training dataset is: ', Y_train[0])
import matplotlib.pyplot as plt
# Thay đô'i kích thước hình mặc định cu'a tâ't ca' các hình vẽ trong
chương trình
plt.rcParams['figure.figsize'] = (9, 9)
# Danh sách các nhãn tương ứng với các loại rác
labels = ['cardboard', 'glass', 'metal', 'paper', 'plastic',
'vegetation']
# Vòng lặp để hiể n thị 25 hình a'nh từ tập huấ n luyện
for i in range(25):
    plt.subplot(5, 5, i + 1) # Tạo 5 hàng và 5 cột cho các hình vẽ
    plt.imshow(X_train[i], interpolation='none') # Hiê'n thị hình
a'nh từ tập huấ n luyên
    plt.title("{}".format(labels[int(Y train[i])])) # Đặt tiêu đê`
cho hình a'nh dưa trên nhãn
plt.tight layout() # Đa'm ba'o các hình không bi chô`ng lên nhau
```

```
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, Conv2D, Flatten
from tensorflow.keras.layers import MaxPooling2D, Activation,
BatchNormalization
from tensorflow.keras.callbacks import TensorBoard, Callback,
EarlyStopping
from tensorflow.keras.optimizers import SGD, RMSprop, Adam, Nadam
from tensorflow.keras.losses import categorical crossentropy
from tensorflow.keras import regularizers
# Sô' lớp phân loai rác: 6 (cardboard, glass, metal, paper, plastic,
vegetation)
num classes = 6
# Xây dưng mô hình CNN
def waste_classification_model():
    model = Sequential()
    # Laver 1: Convolutional laver
    model.add(Conv2D(32, kernel size=(3, 3), activation='relu',
input shape=(50, 50, 3)) # 50 \times 50 là kích thước hình a'nh ban resize
    model.add(MaxPooling2D(pool size=(2, 2)))
    # Layer 2: Convolutional layer
    model.add(Conv2D(64, kernel size=(3, 3), activation='relu'))
    model.add(MaxPooling2D(pool size=(2, 2)))
    model.add(Dropout(0.25))
    # Flatten the layers to connect to fully connected layers
    model.add(Flatten())
    # Fully connected layer
    model.add(Dense(128, activation='relu'))
    model.add(Dropout(0.5))
    # Output layer (6 classes)
    model.add(Dense(num classes, activation='softmax'))
    # Compile the model
    model.compile(optimizer=Adam(learning rate=0.001),
                  loss='sparse categorical crossentropy', # Dùng
sparse categorical crossentropy khi nhãn là số nguyên
                  metrics=['accuracy'])
    model.summary()
    return model
# Khơ'i tạo mô hình
model = waste classification model()
```

```
# Huâ'n luvên mô hình
history = model.fit(X train, Y train, batch size=32, epochs=10,
validation split=0.2)
# Đánh giá mô hình trên tập kiể m thư '
test loss, test acc = model.evaluate(X test, Y test, verbose=2)
print('\nTest accuracy:', test acc)
# Đánh giá trên tập huấ n luyên
train score = model.evaluate(X train, Y train, verbose=0)
print('Train loss:', round(train_score[0], 4))
print('Train accuracy:', round(train score[1], 4), '\n')
# Đánh giá trên tập kiể m thư '
test score = model.evaluate(X test, Y test, verbose=0)
print('Test loss:', round(test score[0], 4))
print('Test accuracy:', round(test_score[1], 4))
from sklearn.metrics import cohen kappa score, confusion matrix,
ConfusionMatrixDisplay
import matplotlib.pyplot as plt
# Dư đoán trên tập kiể m thư '
v pred = model.predict(X test)
y_pred_multiclass = np.argmax(y_pred, axis=1) # Chuyê'n dự đoán thành
nhãn (class)
y test multiclass = Y test # Nhãn thực tê' đã được mã hóa trước đó,
không câ`n dùng argmax
# Cohen Kappa Score
kappa = cohen_kappa_score(y_test_multiclass, y_pred_multiclass)
print("The result of Kappa is:", round(kappa, 3))
# Confusion Matrix (Ma trân nhâ`m lâ~n)
cm = confusion matrix(y test multiclass, y pred multiclass)
display = ConfusionMatrixDisplay(confusion matrix=cm,
display_labels=['cardboard', 'glass', 'metal', 'paper', 'plastic',
'vegetation'l)
# Vẽ biể'u đô` confusion matrix
fig = plt.figure(figsize=(11, 11))
ax = fig.subplots()
display.plot(ax=ax)
plt.show()
```

## 4. Experiments Report

Provide a summary of experimental results, explain the meaning of your result and how your model can be used to address the related business problem.